

COLOR SENSOR

TO ALL WHOM IT MAY CONCERN:

BE IT KNOWN THAT ROGER L. WOODALL, employee of the United States Government, citizen of the United States of America, and resident of Jewett City, County of New London, State of Connecticut, has invented certain new and useful improvements, entitled as set forth above, of which the following is a specification:

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COLOR SENSOR

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STATEMENT OF GOVERNMENT INTEREST

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The invention described herein may be manufactured by or for the Government of the United States of America for Governmental purposes without the payment of any royalties thereon or therefore.

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CROSS-REFERENCE TO RELATED APPLICATIONS

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This patent application is co-pending with related patent applications entitled NEURAL DIRECTORS (U.S. Patent Application Ser. No. 09/436,957), NEURAL SENSORS (U.S. Patent Application Ser. No. 09/436,956), STATIC MEMORY PROCESSOR (U.S. Patent Application Ser. No. 09/477,638), DYNAMIC MEMORY PROCESSOR (U.S. Patent Application Ser. No. 09/477,653), MULTIMODE INVARIANT PROCESSOR (U.S. Patent Application Ser. No. 09/641,395) and A SPATIAL IMAGE PROCESSOR (^{Ser.} ~~Attorney Docket No. 77346~~ ^{09/853,932}), by the same inventor as this patent application.

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BACKGROUND OF THE INVENTION

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(1) Field of the Invention

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The invention relates generally to the field of color sensors and more particularly to color sensors having neural networks with a plurality of hidden layers, or multi-layer neural

1 networks, and further to a new neural network processor for
2 sensing color in optical image data.

3 (2) Description of the Prior Art

4 Electronic neural networks have been developed to rapidly
5 identify patterns in certain types of input data, or accurately
6 to classify the input patterns into one of a plurality of
7 predetermined classifications. For example, neural networks have
8 been developed which can recognize and identify patterns, such as
9 the identification of hand-written alphanumeric characters, in
10 response to input data constituting the pattern of on and off
11 picture elements, or "pixels", representing the images of the
12 characters to be identified. In such a neural network, the pixel
13 pattern is represented by, for example, electrical signals
14 coupled to a plurality of input terminals, which, in turn, are
15 connected to a number of processing nodes, each of which is
16 associated with one of the alphanumeric characters which the)
17 neural network can identify. The input signals from the input
18 terminals are coupled to the processing nodes through certain
19 weighting functions, and each processing node generates an output
20 signal which represents a value that is a non-linear function of
21 the pattern of weighted input signals applied thereto. Based on
22 the values of the weighted pattern of input signals from the
23 input terminals, if the input signals represent a character that
24 can be identified by the neural network, the one of the
25 processing nodes associated with that character will generate a

1 positive output signal, and the others will not. On the other
2 hand, if the input signals do not represent a character that can
3 be identified by the neural network, none of the processing nodes
4 will generate a positive output signal. Neural networks have
5 been developed which can perform similar pattern recognition in a
6 number of diverse areas.

7 The particular patterns that the neural network can identify
8 depend on the weighting functions and the particular connections
9 of the input terminals to the processing nodes. The weighting
10 functions in, for example, the above-described character
11 recognition neural network, essentially will represent the pixel
12 patterns that define each particular character. Typically, each
13 processing node will perform a summation operation in connection
14 with values representing the weighted input signals provided
15 thereto, to generate a sum that represents the likelihood that
16 the character to be identified is the character associated with
17 that processing node. The processing node then applies the non-
18 linear function to that sum to generate a positive output signal
19 if the sum is, for example, above a predetermined threshold
20 value. Conventional non-linear functions which processing nodes
21 may use in connection with the sum of weighted input signals is
22 generally a step function, a threshold function, or a sigmoid, in
23 all cases the output signal from the processing node will
24 approach the same positive output signal asymptotically.

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1 Before a neural network can be useful, the weighting
2 functions for each of the respective input signals must be
3 established. In some cases, the weighting functions can be
4 established a priori. Normally, however, a neural network goes
5 through a training phase, in which input signals representing a
6 number of training patterns for the types of items to be
7 classified, for example, the pixel patterns of the various hand-
8 written characters in the character-recognition example, are
9 applied to the input terminals, and the output signals from the
10 processing nodes are tested. Based on the pattern of output
11 signals from the processing nodes for each training example, the
12 weighting functions are adjusted over a number of trials. After
13 the neural network has been trained, during an operational phase
14 it can generally accurately recognize patterns, with the degree
15 of success based in part on the number of training patterns
16 applied to the neural network during the training stage, and the
17 degree of dissimilarity between patterns to be identified. Such
18 a neural network can also typically identify patterns that are
19 similar, but not necessarily identical, to the training patterns.

20 One of the problems with conventional neural network
21 architectures as described above is that the training
22 methodology, generally known as the "back-propagation" method, is
23 often extremely slow in a number of important applications. In
24 addition, under the back-propagation method, the neural network
25 may result in erroneous results that may require restarting of

1 training. Even after a neural network has been through a
2 training phase, confidence that the best training has been
3 accomplished may sometimes be poor. If a new classification is
4 to be added to a trained neural network, the complete neural
5 network must be retrained. In addition, the weighting functions
6 generated during the training phase often cannot be interpreted
7 in ways that readily provide understanding of what they
8 particularly represent.

9 Edwin H. Land's Retinex theory of color vision is based upon
10 "three color" experiments performed before 1959. A simple
11 "mishap" showed that three colors were not always required to see
12 accurate color. Land used a short and long record of brightness
13 data (black and white transparencies) to produce color perceived
14 by human eyes and not by photographic means. He demonstrated a
15 perception of a full range of pastel colors using two very
16 similar in color light sources such as yellow, at 579 nm and
17 yellow orange, at 599 nm ("Experiments in Color Vision", Edwin H.
18 Land, Scientific American, Vol. 200 No. 5, May 1959). Land found
19 that in some two record experiments all colors present were not
20 perceived. Although Land demonstrated that two records provided
21 color perceptions, he constructed his Retinex theory upon three
22 records such as his long, medium and short records (An
23 Alternative Technique for the Computation of the Designator in
24 the Retinex Theory of Color Vision", Edwin H. Land, Proceedings
25 of the National Academy of Sciences, Vol. 83, 1986). The

1 invention herein is related to human color perception discovered
2 during Land's color vision experiments as reported in 1959.

3 The "Trichromatic" theory in human color vision has been
4 accepted on and off since the time of Thomas Young in 1802 (A
5 Vision in the Brain", S. Zeki, Blackwell Scientific Publishing,
6 1993). Still and video electronic camera designs are correctly
7 based upon the trichromatic theory but the current designs are
8 highly subjective to color error reproduction due to changes in
9 the ambient light color temperatures and color filtrations. The
10 device in this invention senses color using a new "bichromatic"
11 theory, which includes a mechanism that insures color constancy
12 over a large range of ambient color temperatures. The use of two
13 lightness records as used by Land in 1959 is one key to this
14 invention.

15 The bichromatic theory is based upon an interpretation of a
16 biological color process that occurs in the eyes and brain of
17 humans and in some animals. The bichromatic theory is defined as
18 a system that functions together under the following assumptions,
19 accepted principles and rules of procedure, for which FIGS. 4A
20 and 4B are provided for support:

21 (1) The system is a color sensing retina. There are at
22 least two photo transducers in each pixel space in the retina,
23 shown in FIG. 4B as TR(HI) and TR(LO).

24 (2) The two photo transducers sense the color of the light
25 at each pixel's position in a scene of color focused on the

1 retina. Each of the at least photo transducers contains a
2 different spectral response and the wavelength difference between
3 the peaks of a pair of these responses is called the waveband or
4 the spectral bandwidth of the two photo transducers.

5 (3) The two photo transducers have overlapping spectral
6 logarithmic responses where their slopes are opposing each other
7 as indicated in FIG. 4A.

8 (4) The photo transducers have at least two controlled gain
9 amplifiers (CGA) and at least two common controlling circuits.
10 There is one controlled gain amplifier for each photo transducer
11 where each of the at least two common controlling circuits
12 controls the controlled gain amplifiers for all the photo
13 transducers of the same spectral response.

14 (5) The highest energy value in the retina, or the peak
15 energy from a photo transducer of a specific spectral response,
16 controls the output of the common controlling circuits that
17 normalize the logarithmic response of all photo transducers with
18 the same spectral response. Thus, it is always the peak energy
19 photo transducer no matter its position in the retina that
20 controls the common mode gain. The peak response of a photo
21 transducer is relative to the best matched wavelength of energy
22 for all wavelengths of light impinging on the color retina.
23 Therefore, each photo transducer will be continuously normalized
24 to the peak photo transducer signal in response to changes in
25 ambient lighting.

1 (6) In a general discussion herein a normalized photo
2 transducer or a normalized pixel includes the controlled gain
3 amplifier as part of its response. A photo transducer sensing
4 the peak energy or a peak energy sensing photo transducer will
5 only be called as such thus a normalized photo transducer will
6 not specifically include a peak energy sensing photo transducer.

7 (7) There are three color coordinates called hue, lightness
8 and saturation. Three degrees of freedom are required to
9 categorize all combinations of color attributes. Two points in a
10 two dimensional space can be connected by a line. Combinations
11 of positions of these two points in space can produce at least
12 three families of lines in the two dimensional space. The line
13 families are horizontal, vertical and sloped. FIG. 4A shows a
14 two dimensional graph of the responses of two normalized photo
15 transducers. A straight line on the graph may represent the two
16 output values of the normalized photo transducers for a specific
17 input light condition. The graph coordinates are light
18 wavelength for the horizontal axis and signal in a natural log
19 scale for the vertical axis. Output values of the two normalized
20 photo transducers can be represented by three families of lines.

21 (8) The response "curve" of a normalized photo transducer
22 output signal for a normalized light energy input is shown as a
23 straight line, from the maximum response at its wavelength, down
24 to the bottom at the opposite side of the graph. Each response
25 curve of the normalized photo transducers has opposing slopes

1 that cross each other. A normalized photo transducer response
2 over the waveband is given as $TR(b)=ce^{-kx}$, where: x equals the
3 wavelength position in the normalized waveband relative to the
4 maximum response of the photo transducer, i.e., (0 to 1); c, the
5 conversion constant, equals one for a normalized light energy,
6 or, alternately, an integrated CGA value; k equals approximately
7 10; and b is the high or low transducer. The output signal level
8 is symbolized by E1 for the low wavelength normalized photo
9 transducer and E2 for the other.

10 (9) A broad constant energy spectrum of visible light
11 relative to its color temperature "flattens" its spectral energy
12 curve as the color temperature increases from a deep red at 1000
13 °K to a "slightly bluish" white at 10,000 °K. Thus, when the
14 peak energy photo transducers normalize the retina's response,
15 the results are equivalent to "whitening" the pixel's responses
16 in the waveband of sensible colors. In other words, possibly
17 different energies near the wavelength of the maximum
18 sensitivities of the peak energy transducers contain
19 approximately equal spectral energies at the output of the
20 respective controlled gain amplifiers. This process develops a
21 color constancy in ambient lights of different color
22 temperatures.

23 (10) A family of horizontal lines can represent the
24 normalized photo transducer responses to a broadband family of
25 white light from bright through gray to dark. Example 1 on the

1 graph is a representation of this family. A family of vertical
2 lines can represent a family of wavelengths in the waveband.
3 Example 2 on the graph is a representation of the wavelength of a
4 monochromatic light source. Families of sloped lines, from a
5 horizontal position to a vertical position, closely represent a
6 morphing from "white" to a monochromatic light. A change from
7 white light to a light of a pure color is along the axis for the
8 color attribute of saturation. Example 3 on the graph is a
9 representation of a pastel color. The three families of lines
10 are closely mapped to the three color coordinates of hue,
11 lightness and saturation, but not with an exact one to one
12 correlation. A combination of either set of three dimensions of
13 color attributes can be mapped into the other. The two response
14 values of a normalized pixel can represent a line that can move
15 in combinations of the three coordinate ways to represent exact
16 changes in lightness, hue and saturation of colors.

17 (11) The output values of a normalized pixel, in response
18 to a monochromatic light, shall exhibit proportional photo
19 transducer output values of E1 and E2 that are relative to the
20 wave length of the light in the waveband between the two photo
21 transducers. In the case where there is a broad spectrum of
22 light illuminating an object, the different reflective bands of
23 light relative to the wavelength responses of the normalized
24 pixel will produce photo transducer output values in proportional
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1 to values that would be generated by a colored light of the
2 perceived color.

3 (12) Changing the pixel's response from straight lines to
4 curved lines on the logarithmic scale does not change the two
5 point families of lines but it will change the form of the
6 mapping between the two different color attributes.

7 (13) There is another control mode that increases the
8 dynamic range of the sensibility to light of all photo
9 transducers in the retina. This control sums the energy of all
10 spectral responses to adjust an iris to maintain a constant
11 energy to the retina under varying environmental lighting
12 intensities.

13 (14) This bichromatic theory projects that human color
14 vision may not be as commonly believed. The human retina
15 contains three color cones to sense three different wavelengths
16 of light, which may be used as two color pairs such as a
17 blue-green pair and a red-green pair. Each color pair is
18 processed in the visual cortex to map colors that can be
19 associated to the visual space of an object in a scene. The two
20 color pairs and processing will produce a wide range of colors
21 sensed and a wide range of color constancy. Edwin H. Land's
22 pre-1959 experiments using two black and white transparencies and
23 two color filters produced a perception of color. The color
24 perception and constancy occur because the brightest area of one
25 of the projected transparencies normalizes the response of the

1 appropriate set of human color cones to the specific color
2 projected and the same occurs for the other transparency. The
3 normalized human retina now sees varying ratios of brightness
4 (energy) over the visual scene, which produces the perception of
5 colors of light for the specific color temperatures of natural or
6 artificial light. The bichromatic theory of color is an
7 integration of the above fourteen theorems that together define
8 the workings of color perception and color constancy.

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SUMMARY OF THE INVENTION

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It is therefore an object of the invention to provide a new
and improved neural network color sensor.

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It is a further object to provide a neural network color
sensor in which the weighting functions may be determined a
priori.

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Another object of the present invention is to provide a
neural network color sensor, which can be trained with a single
application of an input data set.

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In brief summary, the color sensor generates color
information defining colors of an image, comparison of colors
illuminated under two or more light sources and boundaries
between different colors. The color sensor includes an input
section, a color processing section, a color comparison section,
a color boundary processing section and a memory processing
section. The input section includes an array of transducer

1 pairs, each transducer pair defining one of a plurality of pixels
2 of the input section. Each transducer pair comprises at least
3 two transducers, each generating an output having a peak at a
4 selected color, the selected color differing as between the two
5 transducers, and each transducer having an output profile
6 comprising a selected function of color. The color processing
7 section includes a plurality of color pixel processors, each
8 receiving the outputs from the two transducers comprising the
9 transducer pair associated with a pixel. In response, the color
10 processing section generates a color feature vector
11 representative of the brightness of the light incident on the
12 pixel and a color value corresponding to the ratio of outputs
13 from the transducers comprising the transducer pair associated
14 with the pixel. The color boundary processing section generates
15 a plurality of color boundary feature vectors, each associated
16 with a pixel, each representing the difference between the color
17 value generated by the pixel color processor for the respective
18 pixel and color values generated by the pixel color processor for
19 pixels neighboring the respective pixel.

20 The color boundary sensor produces object shape feature
21 vectors from a function of the differences in color. This color
22 boundary sensor can sense a colored object shape in a color
23 background where a black and white sensing retina could not
24 detect differences in lightness between the background and the
25 object. The color comparator processor can measure and compare

1 the reflective color of two objects, even when each object is
2 illuminated by two lights of different color temperatures.
3 Thememory processor section provides a process to recognize a
4 color, a boundary of color and a comparison of colors.

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BRIEF DESCRIPTION OF THE DRAWINGS

7 A more complete understanding of the invention and many of
8 the attendant advantages thereto will be readily appreciated as
9 the same becomes better understood by reference to the following
10 detailed description when considered in conjunction with the
11 accompanying drawings wherein corresponding reference characters
12 indicate corresponding parts throughout the several views of the
13 drawings and wherein:

14 FIG. 1 is a functional block diagram of a color sensor
15 constructed in accordance with the invention;

16 FIG. 1A is an expanded view of a transducer pair;

17 FIG. 2 is a functional block diagram of a color processor,
18 which is useful in the color sensor depicted in FIG. 1;

19 FIG. 3 is a functional block diagram of a color boundary
20 processor, which is useful in the color sensor of in FIG. 1;

21 FIG. 4A is an example of the responses of two normalized
22 photo transducers used in the color sensor; and

23 FIG. 4B is a schematic illustration of the theorems defining
24 the workings of the color sensor.

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DESCRIPTION OF THE PREFERRED EMBODIMENT

FIG. 1 is a functional block diagram of a color sensor 10 constructed in accordance with the invention. By way of background, the color sensor 10 operates in accordance with a "bi-chromatic" mechanism of color recognition, which is theorized as being similar to the way in which human beings see and recognize color. In the conventional "tri-chromatic" color recognition mechanism, any color light, either reflected or incidental, can be generated combining three different color illuminations. In the reverse, i.e., color recognition, any input color can be represented or analyzed as a combination of three colors, i.e., base colors. Accordingly three transducers, each sensing one of the base three colors, can be used to determine the contribution of each of the base colors in the input color. In the bi-chromatic mechanism, colors can be distinguished using two color transducers, which have peak sensitivity at different colors, and provide a known output signal response as a function of the input color. The color sensor 10 determines, for an input image, the distribution of colors over the image, using two color transducers to identify the color at each point (that is, for each pixel or picture element) in the image. The color boundary process produces object shape features relative to the boundaries between different colors. The color comparator process produces comparative features relative to a "true reflective color" in

1 ambient lights of different color temperatures. The reading of a
2 "true reflective color" in an ambient light of a color
3 temperature and the reading of the same "true reflective color"
4 in an ambient light of a second color temperature is a process
5 that mimics human color constancy.

6 With reference to FIG. 1, the color sensor 10 includes an
7 input section 11, a color processing section 12 and a color
8 boundary processing section 13, a color comparison processor 19
9 and a memory processor 29. The color processing section 12 and a
10 color boundary processing section 13 both generate color and
11 color boundary feature vectors, which may be provided to, for
12 example, a memory processing section 14. The input section 11
13 receives an image of an object and generates, for each point, or
14 pixel, color information signals representative of the color at
15 the particular point of the image. The input section 11 includes
16 a "retina" 15, which comprises an array of transducer pairs 15(1)
17 through 15(M) (generally identified by reference numeral 15(m)
18 and shown in the expanded view of FIG. 1A), which define the
19 pixels of the image. Each transducer pair comprises two
20 transducers, which have output peaks at two different frequencies
21 and which provide a predetermined output value as a function of a
22 color wave band. Preferably, all of the pixels will have one
23 transducer 15(m) (1) which has a peak output at one frequency
24 identified as 1 and the second transducer 15(m) (2) having a peak
25 output at a second frequency identified as 2. The input section

1 11 further includes a lens 26, which focuses an image of the
2 object onto the retina 15, and an iris 17, which controls the
3 intensity of light incident on the retina 15.

4 The color processing section 12 uses the color information
5 signals from the input section to generate, for each pixel, a
6 local color feature vector representative of the color of the
7 pixel. The color processing section 12 consists of a color
8 processor array 20 and a feature fusion network array 23. The
9 structure and operation of the color processing section 12 will
10 be described in detail below in connection with FIG. 2.
11 Similarly, the color boundary processing section 13 generates,
12 for each pixel, a local color gradient feature vector that
13 represents the gradient of the color at the pixel. The structure
14 and operation of the color boundary processing section 13 will be
15 described in detail below in connection with FIG. 3. The memory
16 processor 29 is as described in STATIC MEMORY PROCESSOR, U.S.
17 Patent Application Ser. No. 09/477,638. The parallel memory
18 processors 16 and 18 are as described for the memory processor of
19 the MULTIMODE INVARIANT PROCESSOR (U.S. Patent Application Ser.
20 No. 09/641,395). The multi-mode invariant image processor,
21 without its input sensor, is used for both parallel memory
22 processors 16 and 18. The possible multiple outputs of the
23 parallel memory processor 18 are the colored input object(s)
24 classifications. The output vector array of the parallel memory
25 processor 16 is a Positional King Of the Mountain (PKOM) array

1 mapped to the pixels 15(m) in the retina, which becomes a map of
2 color classifications of each pixel. It is noted that the PKOM
3 array is a neural network array internal to the parallel memory
4 processor 16 and the remaining neural circuits to the normal
5 output of the MULTIMODE INVARIANT PROCESSOR are not used. The
6 memory processor 29 is a static memory processor and provides an
7 output classification as a degree of color comparison.

8 The local color feature vectors and the local color gradient
9 feature vectors generated for all of the pixels are processed by
10 the processing section 14 to, for example, classify the image
11 into one of a plurality of image classes. The processing section
12 14 may comprise any of a plurality of processing elements for
13 processing the vectors generated by the color processors 12, 13
14 and/or 19.

15 FIG. 2 is a functional block diagram of color processing
16 section 12 and 19 as used in the color sensor of FIG. 1. With
17 reference to FIG. 2, the color processing section 12 includes a
18 plurality of pixel color processors 20(1) through 20(M),
19 generally identified by reference numeral 20(m). For each color
20 processor 20(m), a corresponding feature fusion network 23(m) of
21 color processing section 12 includes corresponding feature fusion
22 neural directors 35(1) through 35(M) and Multi King Of the
23 Mountain (MKOM) 36(1) through 36(M), generally identified by
24 reference numerals 35(m) and 36(m), respectively. The structures
25 of all of the pixel color processors 20(m) are similar, and so

1 FIG. 2 depicts the structure of only one pixel color processor
2 and the corresponding feature fusion neural director 35(m) and
3 MKOM 36(m). Each pixel color processor 20(m) processes the
4 outputs generated by one of the transducer pairs in the retina
5 11. The color processing section 12 also includes a common
6 control 21, which controls all of the pixel color processors
7 20(m) in parallel, controls the iris 17 and receives pixel data
8 from each color processor 20(m).

9 Each pixel color processor 20(m) includes controlled gain
10 amplifier (CGA) circuits 30(m)(1), 30(m)(2), which receive the
11 color amplitude signals generated by the respective transducers
12 15(m)(1), 15(m)(2). Each CGA circuit 30(m)(1), 30(m)(2)
13 generates an output adjusted by a gain control factor generated
14 by the common control 21. The gain control factor is a function
15 of the output of the transducer for each frequency having the
16 highest amplitude, referred to as 15(H)(1) and 15(H)(2). The CGA
17 circuits 30(m)(1), 30(m)(2) will normalize the respective outputs
18 in relation to the highest amplitude output for their respective
19 frequency. This allows each transducer pair 15(m) and their
20 respective CGA circuit 30(m) to output differing values, which
21 represent the color at each transducer pair 15(m) as well as the
22 "color temperature" of the light incident on the object or retina
23 15. The common control 21 senses all transducer outputs for each
24 frequency and uses the highest outputs 15(H)(1), 15(H)(2) to set
25 each CGA circuit 30(m) in the color processor 12 to the same gain

1 as the CGA circuits 30(H)(1), 30(H)(2) from the pixel(s) 15(m)
2 that sensed the highest light energy in retina 15. The
3 transducers 15(H)(1), 15(H)(2), the CGA circuits 30(H)(1),
4 30(H)(2) and the common control 21 operate as an automatic gain
5 controlled loop normalizing the output signal at CGA circuit
6 30(H)(1). Therefore, the response of each transducer 15(m)(1) is
7 normalized at the output of each CGA circuit 30(m)(1) relative to
8 the output of CGA circuit 30(H)(1). It is to be noted that the
9 transducers 15(H)(1), 15(H)(2) need not be from the same pixel
10 15(m), as the spectral light energy of a visual scene image at
11 two separate frequencies is generally not the same everywhere on
12 retina 15.

13 The gain controlled output of each CGA circuit 30(m)(1),
14 30(m)(2) is provided to a number of elements, including a
15 respective sum circuit 33(m), a difference circuit 32(m) and the
16 common control 21. The outputs from the CGA circuits 30(m)(1),
17 30(m)(2) are coupled to the difference circuit, or difference
18 generator 32(m), which generates an output vector that is
19 representative of the difference between the amplitudes of the
20 outputs from the CGA circuits 30(m)(1), 30(m)(2). Accordingly,
21 it will be appreciated that the output generated by the
22 difference generator 32(m) corresponds to the ratio of the
23 amplitudes of the automatic controlled gain signals from the
24 respective transducers 15(H)(1), 15(H)(2) and the respective
25 pixel transducer 15(m) outputs.

1 As noted above, the outputs from the CGA circuits 30(m) (1)
2 and 30(m) (2) are also coupled to a sum circuit 33(m). The sum
3 circuit 33(m) generates an output that corresponds to the sum of
4 the amplitudes of the automatic controlled gain signal from the
5 respective transducers 15(m) (1) and 15(m) (2), and thus represents
6 the brightness of the light incident on the pixel defined by the
7 transducers.

8 The output vector from difference circuit 32(m) is coupled
9 to the color boundary processor 13 (FIG. 1). The difference
10 vector from difference circuit 32(m) and the brightness vector
11 from sum circuit 33(m) are also both coupled to a neural director
12 35(m) that disperses these inputs into a local color feature
13 vector. The neural director 35(m) is preferably similar to the
14 neural directors as described in NEURAL DIRECTOR, U.S. Patent
15 Application Ser. No. 09/436,957. Neural director 35(m) is
16 preferably established to provide an output vector with an
17 increased dimensionality, which will aid in distinguishing
18 between similar patterns in the input vector.

19 The output of the neural director 35(m) is coupled to bi-
20 polar MKOM 36(m), which is described in detail in STATIC MEMORY
21 PROCESSOR, U.S. Patent Application Ser. No. 09/477,638. The bi-
22 polar MKOM 36(m) generates a number of positive and/or negative
23 outputs M(1) through M(R), generally identified by reference
24 numeral M(r), each of which is associated with one dimension of
25 the feature vector input thereto. Each positive component M(r)

1 of the output vector can have a range of values from zero up to a
2 maximum value, which corresponds to, or is proportional to, the
3 maximum positive element value of the input vector. The positive
4 outputs $M(r)$ that are associated with an input vector component
5 having successively lower positive values, are themselves
6 successively lower in value, thus forming a positive ranking of
7 the vector components. Outputs $M(r)$ that are associated with
8 input vector components having negative values are also ranked as
9 negative vector components in a similar manner to the positive
10 components. The rankings for the respective input feature
11 vectors may be global, for all of the components of the input
12 feature vector, or they may be localized among a selected number
13 of preferably contiguous input feature vector components. The
14 feature vector generated by the bi-polar MKOM 36(m) is coupled to
15 the memory processing section 14.

16 The outputs from CGA circuits 30(m)(1) and 30(m)(2) of all
17 of the pixel color processors 20(m) are also coupled to the
18 common control 21. The common control 21 includes peak sensing
19 circuits 40(1), 40(2), each of which receives the output from the
20 correspondingly-indexed CGA circuits 30(m)(1), 30(m)(2), and each
21 generates an output which corresponds to the one of the outputs
22 from the correspondingly-indexed CGA circuits 30(m)(1), 30(m)(2)
23 with the largest signal value. The outputs from the peak
24 circuits 40(1), 40(2) are also connected to control the gain of
25

1 all of the correspondingly-indexed CGA circuits 30(m) (1),
2 30(m) (2) .

3 The outputs from the CGA circuits 30(m) (1) and 30(m) (2) of
4 all of the color pixel processors 20(m) are also connected to a
5 sum circuit 41. The sum circuit 41 generates an output, which
6 represents the sum of the outputs from all of the CGA circuits
7 30(m) (1), 30(m) (2) of all of the color pixel processors 20(m) .
8 The output provided by the sum circuit 41 represents the total
9 intensity or power of the light incident on the retina 15. An
10 iris control circuit 42 uses the sum circuit 41 output to control
11 the iris 17, which normalizes the intensity of the light on
12 retina 15.

13 FIG. 3 is a functional block diagram of the color boundary
14 processor 13, which is useful in the color sensor depicted in
15 FIG. 1. The color boundary processor 13 can sense a colored
16 object shape in a background of a different color. A black and
17 white sensing retina often responds to different colors as equal
18 lightness. Therefore, it may not sense an object of one color
19 against a different background color. As noted above, the color
20 boundary processor 13 receives the color vector signals from the
21 difference circuits 32(m) of all of the pixel color processors
22 20(m) . Color boundary processor 13 then generates an output for
23 each pixel 15(m) that represents a color gradient for the pixel
24 15(m) . The outputs of each difference circuit 32(m) are
25 spatially arranged exactly in the same spatial orientation as

1 each associated pixel 15(m) in retina 15. The array of
2 difference circuit 32(m) outputs becomes a virtual retina 55,
3 shown in FIG. 3 to aid in the visualization of the spatial
4 interconnections between the array of color processors 20 and
5 color boundary processor 13. The color boundary processor 13
6 comprises a plurality of window difference networks 50(1) through
7 50(M), generally identified by reference numeral 50(m), each
8 associated with one of the pixels 15(m) and associated window
9 57(m). Color boundary processor 13 further comprises a like
10 plurality of neural directors 51(m).

11 Each window difference network 50(m) receives a local window
12 array 57(m) of difference vectors generated by the
13 correspondingly-indexed pixel color processor 20(m). Each window
14 difference network 50(m), in turn, generates an output vector
15 which represents a color acceleration vector between the
16 difference vectors provided by the correspondingly-indexed pixel
17 color processor 20(m) and color vectors for pixels within a
18 predetermined area around the pixel 15(m), illustrated in FIG. 3
19 as local window 57(m). Local window 57(m) may consist of any
20 chosen pattern of pixels surrounding pixel 15(m). e.g., a star
21 pattern or a box pattern. Each neural director 51(m) receives
22 the color acceleration vector from the correspondingly-indexed
23 window difference network 50(m). As with neural director 35(m),
24 each neural director 51(m) is preferably established to provide
25 an output local color boundary feature vector with the same or an

1 increased dimensionality, which will aid in distinguishing
2 between similar patterns in the input vector.

3 In a modification to the invention 10, each pixel can be a
4 three transducer set 15(m). Each transducer of the set 15(m) is
5 to be matched to the response of the human retinal color cones.
6 The three transducer set 15(m) will produce two "transducer
7 pairs" for each pixel 15(m) and with two color processors 12 a
8 color retina will be produced. The retina and two parallel memory
9 processors 16 will sense color matched to the human color
10 perception over a wide range of ambient lighting conditions.

11 With reference again to FIG. 1, the local color feature
12 vectors generated by the pixel color processing section 12, an
13 array of color comparators 19 and the local color boundary
14 feature vectors generated by color boundary processor 13 for all
15 of the pixels 15(m), are coupled to the memory processing section
16 14. The memory processing section 14 may perform a variety of
17 individual or combined operations in connection with the feature
18 vectors input thereto, including object recognition and the like,
19 based on preselected object classification patterns or the like.

20 The invention provides a number of advantages. In
21 particular, the invention provides a system for receiving an
22 image of an object and generates, for an array of pixels of the
23 image, color and color gradient/boundary information, in the form
24 of feature vectors, which may be processed to, for example,
25 classify the object into one of a plurality of object classes.

1 The system generates the color and color gradient/boundary
2 information using only two transducers for each pixel, in
3 accordance with a bi-chromatic color recognition scheme, with the
4 transducers having peak responses at selected colors 1 and 2,
5 and a known output profile as a function of color, instead of the
6 non-color constancy process produced in accordance with the tri-
7 chromatic color recognition scheme.

8 It will be appreciated that numerous modifications may be
9 made to the system 10. For example, the memory processing
10 section 14 may perform processing in connection with comparisons
11 generated for two images, using output color feature vectors
12 generated either by the same color sensor 10 at two points in
13 time, or output comparator vectors which are generated by two
14 color sensors (the second being denoted by 11' and 12') for
15 respective pixels 15(m) for respective images. In that case, and
16 with reference to FIG. 2, the color processing section 12, in
17 particular the pixel color processors 20(m), may provide outputs
18 for the two images to the respective difference circuits 60(m),
19 61(m) of color comparison processor 19, each of which generates a
20 difference vector representing the difference between the
21 difference vectors and brightness vectors generated by the color
22 processors 12 for the respective images. The difference vectors
23 of 60(m) and 61(m) are input to comparator feature fusion network
24 array 62, which operates in a manner similar to feature fusion
25 network array 23. Similar difference circuits (not shown) may

1 also be provided for the local color boundary feature vectors
2 generated by the color difference processors 13 for the
3 respective images.

4 In addition, the peak detector circuits 40(1), 40(2) of the
5 common control 21 may be replaced with summing circuits that
6 generate a sum output for controlling the CGA circuits 30(m)(1),
7 30(m)(2).

8 Preferably, the iris control 42 will generally rapidly
9 adjust the iris in response to changes in the light intensity
10 levels incident on the retina 15, so as to maintain the light
11 levels incident on the transducers within a predetermined
12 operating range. In that case, the CGA circuits 30(m)(1),
13 30(m)(2) may have a relatively slower response to changes in the
14 automatic gain control signals from the control circuit 21.
15 These differences in response will allow the slower response of
16 normalization via the CGA circuits to maintain a steady color
17 constancy in a scene of rapid brightness changes.

18 The described components of invention 10 provide the
19 necessary components for a uniquely designed photographer's
20 exposure and color temperature meter. A calibration of the
21 common control network 21 provides values for exposure and color
22 temperature data. The meter may be an independent device, i.e.,
23 a hand held meter, or it may be integrated in a camera body,
24 either electronic or film, to provide automatic exposure and
25

1 color temperature corrections. The device may also be integrated
2 into color printers or printing presses as a color ink control.

3 It will be apparent that variations and modifications may be
4 made to the invention herein described and illustrated, by those
5 skilled in the art with the attainment of some or all of the
6 advantages of the invention. It is also understood that the
7 color sensor described herein may be connected to the various
8 devices described in the referenced patent applications, wherein
9 all the devices act in concert in a manner similar to the human
10 eye. Therefore, it is the object of the appended claims to cover
11 all such variations and modifications as come within the true
12 spirit and scope of the invention.